

Machine Learning 1

Lecture 1.5 - Probability Theory: An Example

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(Bishop 1.2.0 - 1.2.1)



Example: Fruit in Boxes

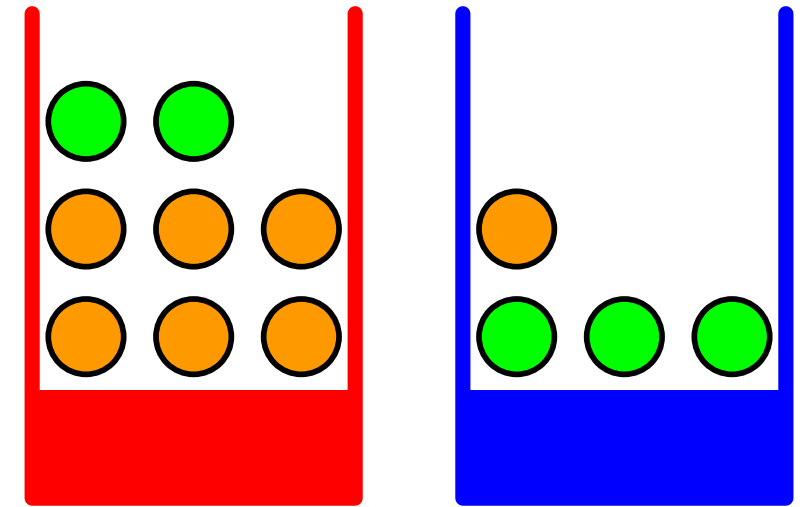


Figure: coloured boxes containing apples and oranges (Bishop 1.9)

Random variables:
 fruit: $F = \{\text{apple}(a), \text{orange}(o)\}$
 box: $B = \{\text{red}(r), \text{blue}(b)\}$

- Prior Box distribution:

$$p(B = r) = 4/10$$

$$p(B = b) = 6/10$$

- Conditional probabilities of Fruit given Box

$$P(F=o | B=b) = \frac{1}{4}$$

$$P(F=o | B=r) = \frac{6}{8} = \frac{3}{4}$$

$$P(F=a | B=b) = \frac{3}{4}$$

$$P(F=a | B=r) = \frac{1}{4}$$

- Marginal Fruit distributions:

$$p(F = a) = \sum_B p(F=a, B) = \sum_B p(F=a | B) \cdot p(B) = \frac{3}{4} \cdot \frac{6}{10} + \frac{1}{4} \cdot \frac{4}{10} = \frac{11}{20}$$

$$p(F = o) = p(F=o | B=r) p(B=r) + p(F=o | B=b) p(B=b) = \frac{3}{4} \cdot \frac{4}{10} + \frac{1}{4} \cdot \frac{6}{10} = \frac{9}{20} \equiv \left(1 - \frac{11}{20}\right) = \frac{9}{20}$$

Example: Fruit in Boxes

- ▶ Prior: $p(B = r) = 4/10$ & $p(B = b) = 6/10$
- ▶ Marginal: $p(F = a) = 11/20$ & $p(F = o) = 9/20$
- ▶ Posterior probability of Box color given observed fruit

$$p(B = r | F = o) = \frac{P(F=o | B=r) \cdot P(B=r)}{P(F=o)}$$

$$= \frac{6}{9} \cdot \frac{4}{10} \cdot \frac{20}{9} = \frac{6}{9} = \frac{2}{3} \approx 66\%$$

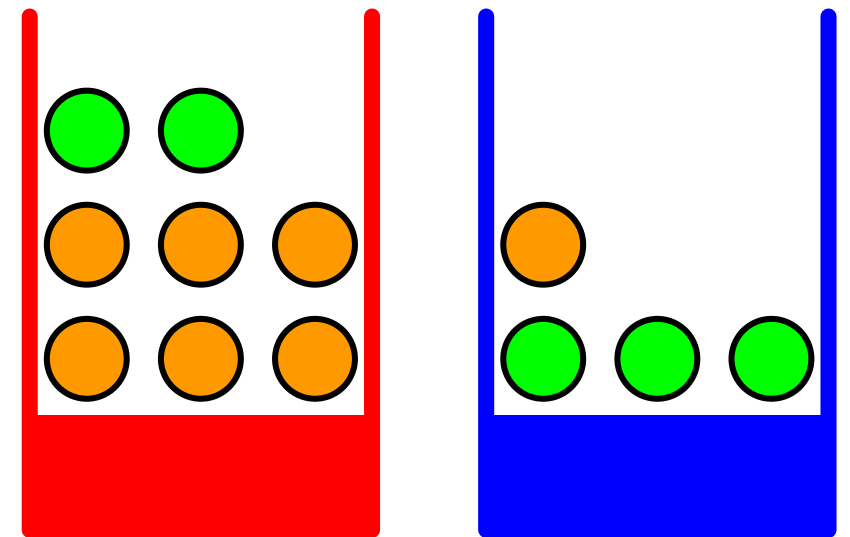


Figure: coloured boxes containing apples and oranges (Bishop 1.9)

- ▶ prior probability of red box:

$$p(B = r) = 4/10 < P(B=r | F:o)$$

- ▶ After observing an orange the probability of observing a red box is now larger than observing a blue box!

Independent Random Variables

Two random variables X and Y are *independent* iff measuring X gives no information on Y , and vice versa.

- Formally: X and Y are called independent if

$$P(x, y) = P(x) P(y)$$

for all $x \in \mathcal{X}, y \in \mathcal{Y}$

$$P(x|y) \cdot P(y)$$

- Equivalent to

$$p(x|y) = \frac{p(x, y)}{p(y)} = \frac{p(x) p(y)}{p(y)} = p(x)$$

- Example:

$$P(F|B) = P(F) = \frac{1}{2}$$

$$P(B|F) = P(B)$$

